```
from sklearn. datasets import fetch_openml
X, y = fetch_openml('mnist_784', version=1, return_X_y=True)
#rescale
X = X / 255.
X_{train}, X_{test} = X[:60000], X[60000:]
y_{train}, y_{test} = y[:60000], y[60000:]
import numpy as np
from sklearn. decomposition import PCA
import matplotlib.pyplot as plt
pca = PCA()
X_train. shape
      (60000, 784)
print(X_train[0])
       0. 32156863 0. 21960784 0. 15294118 0.
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```

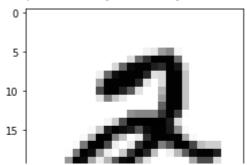
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n_{comp} = 30
pca = PCA (n_components=n_comp)
pca. fit(X_train)
#30次元の潜在空間に変換して圧縮
X_train_latent = pca. transform(X_train)
X_train_latent.shape
      (60000, 30)
X_train_latent[0]
      array ([ 0.48601015, -1.22617358, -0.09613354, -2.17944298, -0.10704577,
              -0. 91167171,   0. 91763043,   0. 62666548,  -1. 4255497 ,
                                                                         0.77815136,
               0. 77449643, -0. 99630654, -0. 44509738, 2. 93841289,
                                                                         0.85980051,
              -0. 01848352.
                             1. 29452532,
                                           1. 21255137,
                                                          1.08903686,
                                                                         0.65102536,
               0. 10739394, -0. 25150383, -0. 8499285,
                                                          0.98205115,
                                                                         0. 1859757
               0.40418704, -1.22074111, 0.65974312, -0.49827378, -0.55485382])
import numpy as np
import matplotlib.pyplot as plt
import matplotlib. font manager
from sklearn import svm
# fit the model
clf = svm. OneClassSVM(nu=0.2, kernel='rbf', gamma=0.1)
clf. fit(X_train_latent)
```

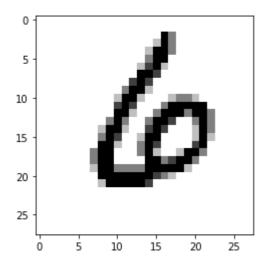
OneClassSVM(cache_size=200, coef0=0.0, degree=3, gamma=0.1, kernel='rbf',

```
max_iter=-1, nu=0.2, shrinking=True, tol=0.001, verbose=False)
y_pred_train_latent = clf.predict(X_train_latent)
print(y_pred_train_latent)
      [1 1 1 ... -1 1 1]
y_pred_train_latent.shape
      (60000,)
y_pred_train_latent[0]
     1
extension_2 =[]
for i in range (10):
    if i%2==0:
        extension_2.append(i)
extension_2
#>>> [0, 2, 4, 6, 8]
     [0, 2, 4, 6, 8]
kara = []
for i in range (100):
   if y_pred_train_latent[i]==-1:
              kara.append(i)
kara
      [27, 28, 31, 34, 47, 51, 56, 60, 63, 64, 70, 75, 82, 83]
len (kara)
     14
kara[0]
     27
plt. imshow(X[5].reshape(28, 28), cmap=plt.cm.gray_r)
```





for i in range(len(kara)):
 plt.imshow(X[kara[i]].reshape(28, 28), cmap=plt.cm.gray_r)



import numpy as np import cv2

from google.colab import drive
drive.mount('/content/gdrive')

Mounted at /content/gdrive

%cd ./gdrive/'My Drive'/"Colab Notebooks"/"extractdouga"

/content/gdrive/My Drive/Colab Notebooks/extractdouga

cv2. imwrite('henkantest.jpg', 255**X[kara[0]].reshape(28, 28, 1))

True

X[kara[0]]. reshape (28, 28, 1)

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```
for i in range(len(kara)):
    cv2. imwrite(f' comb {i:03d}. jpg', 255**X[kara[i]]. reshape(28, 28, 1))
```

#ファイルに問題あり

```
import sys
import cv2
```

```
# encoder (for mp4)
fourcc = cv2. VideoWriter_fourcc('m', 'p', '4', 'v')
# output file name, encoder, fps, size(fit to image size)
video = cv2. VideoWriter('video.mp4', fourcc, 20.0, (1240, 1360))
if not video.isOpened():
    print("can't be opened")
    sys.exit()
for i in range (0, len (kara) ):
    # hoge0000.png, hoge0001.png,..., hoge0090.png
    img = cv2. imread('./comb%03d.jpg' % i)
    # can't read image, escape
    if img is None:
        print("can't read")
        break
    # add
    video.write(img)
    print(i)
video.release()
print('written')
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 C→
     2
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     6
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     8
     10
     11
     12
     13
     written
import glob
from PIL import Image
frames = []
images = sorted(glob.glob("./*.jpg"))
print(images)
      ['./comb000.jpg', './comb001.jpg', './comb002.jpg', './comb003.jpg', './comb004.jpg', './comb
for image in images:
```

✓ 0秒 完了時間: 13:46

×